# Stochastic Uncoupled Dynamics and Nash Equilibrium\*

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#### Abstract

In this paper we consider dynamic processes, in repeated games, that are subject to the natural informational restriction of uncoupledness. We study the almost sure convergence to Nash equilibria, and present a number of possibility and impossibility results. Basically, we show that if in addition to random moves some recall is introduced, then successful search procedures that are uncoupled can be devised. In particular, to get almost sure convergence to pure Nash equilibria when these exist, it suffices to recall the last two periods of play.

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#### 1 Introduction

A dynamic process in a multi-player setup is *uncoupled* if the moves of every player do not depend on the payoff (or utility) functions of the other players.

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This is a natural informational requirement, which holds in most models. In Hart and Mas-Colell (2003) we introduce this concept and show that uncoupled stationary dynamics cannot always converge to Nash equilibria, even if these exist and are unique. The setup was that of deterministic, stationary, continuous-time dynamics.

It is fairly clear that the situation may be different when *stochastic* moves are allowed, since one may then try to carry out some version of exhaustive search: keep randomizing until by pure chance a Nash equilibrium is hit, and then stop there. However, this is not so simple: play has a decentralized character, and no player can, alone, recognize a Nash equilibrium. The purpose of this paper is, precisely, to investigate to what extent Nash equilibria can be reached when considering dynamics that satisfy the restrictions of our previous paper: uncoupledness and stationarity. As we shall see, in addition to making random moves, any positive result will also require the players to recall pieces of the past.

Because we allow random moves, it is easier to place ourselves in a discrete time framework. Thus we consider the repeated play of a given stage game, under the standard assumption that each player observes the play of all players; as for payoffs, each player knows only his own payoff function. We start by studying a natural analog of the approach of our earlier paper; that is, we assume that in determining the random play at time t+1 the players retain only the information contained in the current play of all players at time t; i.e., past history does not matter. We call this the case of 1-recall. We shall then see that the result of our earlier paper is recovered: convergence to Nash equilibrium cannot be ensured under the hypotheses of uncoupledness, stationarity, and 1-recall (there is an exception for the case of generic two-player games with at least one pure Nash equilibrium).

Yet, the exhaustive search intuition can be substantiated if we allow for (uncoupled and stationary) strategies with longer recall. Perhaps surprisingly, to guarantee almost sure convergence to pure Nash equilibria when these exist, it suffices to have 2-recall: to determine the play at t+1 the players use the information contained in the plays of all the players at periods t and t-1. In general, when Nash equilibria may be mixed, we show that

convergence to (approximate) equilibria can be guaranteed using longer, but finite, recall.

Finally, one can view this paper as contributing to the demarcation of the border between those classes of dynamics for which convergence to Nash equilibrium can be obtained and those for which it cannot.

The paper is organized as follows. Section 2 presents the model and defines the relevant concepts. Convergence to pure Nash equilibria is studied in Section 3, and to mixed equilibria, in Section 4 (with a proof relegated to the Appendix). We conclude in Section 5 with some comments and a discussion of the related literature, especially the work of Foster and Young (2003a, 2003b).

# 2 The Setting

A basic static (one-shot) game is given in strategic (or normal) form, as follows. There are  $N \geq 2$  players, denoted i = 1, 2, ..., N. Each player i has a finite set of actions  $A^i$ ; let  $A := A^1 \times A^2 \times ... \times A^N$  be the set of action combinations. The payoff function (or utility function) of player i is a real-valued function  $u^i : A \to \mathbb{R}$ . The set of randomized or mixed actions of player i is the probability simplex over  $A^i$ , i.e,  $\Delta(A^i) = \{x^i = (x^i(a^i))_{a^i \in A^i} : \sum_{a^i \in A^i} x^i(a^i) = 1 \text{ and } x^i(a^i) \geq 0 \text{ for all } a^i \in A^i\}$ ; as usual, the payoff functions  $u^i$  are multilinearly extended, so  $u^i : \Delta(A^1) \times \Delta(A^2) \times ... \times \Delta(A^N) \to \mathbb{R}$ .

We fix the set of players N and the action sets  $A^i$ , and identify a game by its payoff functions  $U = (u^1, u^2, ..., u^N)$ .

For  $\varepsilon \geq 0$ , a Nash  $\varepsilon$ -equilibrium x is an N-tuple of mixed actions  $x = (x^1, x^2, ..., x^N) \in \Delta(A^1) \times \Delta(A^2) \times ... \times \Delta(A^N)$ , such that  $x^i$  is an  $\varepsilon$ -best reply to  $x^{-i}$  for all i; i.e.,  $u^i(x) \geq u^i(y^i, x^{-i}) - \varepsilon$  for every  $y^i \in \Delta(A^i)$  (we write  $x^{-i} = (x^1, ..., x^{i-1}, x^{i+1}, ..., x^N)$  for the combination of mixed actions of all players except i). When  $\varepsilon = 0$  this is a Nash equilibrium, and when  $\varepsilon > 0$ , a Nash approximate equilibrium.

The dynamic setup consists of a repeated play, at discrete time periods t = 1, 2, ..., of the static game U. Let  $a^{i}(t) \in A^{i}$  denote the action of player

i at time<sup>1</sup> t, and put  $a(t) = (a^1(t), a^2(t), ..., a^N(t)) \in A$  for the combination of actions at t. We assume that there is standard monitoring: at the end of period t each player i observes everyone's realized action, i.e., a(t).

A strategy<sup>2</sup>  $f^i$  of player i is a sequence of functions  $(f_1^i, f_2^i, ..., f_t^i, ...)$ , where, for each time t, the function  $f_t^i$  assigns a mixed action in  $\Delta(A^i)$  to each history (a(1), a(2), ..., a(t-1)). A strategy profile is  $f = (f^1, f^2, ..., f^N)$ .

A strategy  $f^i$  of player i has finite recall if there exists a positive integer R such that only the history of the last R periods matters: for each t > R, the function  $f^i_t$  is of the form  $f^i_t(a(t-R), a(t-R+1), ..., a(t-1))$ ; we call this R-recall. Such a strategy is moreover stationary if the ("calendar") time t does not matter:  $f^i_t \equiv f^i(a(t-R), a(t-R+1), ..., a(t-1))$  for all t > R.

Strategies have to fit the game being played. We thus consider a strategy mapping, which, to every game (with payoff functions) U, associates a strategy profile  $f(U) = (f^1(U), f^2(U), ..., f^N(U))$  for the repeated game induced by  $U = (u^1, u^2, ..., u^N)$ . Our basic requirement for a strategy mapping is uncoupledness, which says that the strategy of each player i may depend only on the i-th component  $u^i$  of U, i.e.,  $f^i(U) \equiv f^i(u^i)$ . Thus, for any player i and time t, the strategy  $f^i_t$  has the form  $f^i_t(a(1), a(2), ..., a(t-1); u^i)$ . Finally, we will say that a strategy mapping has k-recall and is stationary if, for any k, the strategies k0 of all players k1 have k2-recall and are stationary.

# 3 Pure Equilibria

We start by considering games that possess pure Nash equilibria (i.e., Nash equilibria  $x = (x^1, x^2, ..., x^N)$  where each  $x^i$  is a pure action in  $A^i$ ). Our first result generalizes the conclusion of Hart and Mas-Colell (2003). We show that with 1-recall — that is, if actions depend only on the current play and not on past history — we cannot hope in all generality to converge, in an uncoupled and stationary manner, to pure Nash equilibria when these exist.

<sup>&</sup>lt;sup>1</sup>More precisely, the actual *realized* action (when randomizations are used).

<sup>&</sup>lt;sup>2</sup>We use the term "strategy" for the repeated game and "action" for the one-shot game.

<sup>&</sup>lt;sup>3</sup>From now on, "game" and "equilibrium" will always refer to the one-shot stage game.

**Theorem 1** There are no uncoupled, 1-recall, stationary strategy mappings that guarantee almost sure convergence to pure Nash equilibria in all games where such equilibria exist.

**Proof.** The following two examples, the first with N=2 and the second with N=3, establish our result. We point out that the second example is generic — in the sense that the best reply is always unique — while the first is not; this will matter in the sequel.

The first example is the two-player game of Figure 1. The only pure Nash equilibrium is  $(\gamma, \gamma)$ . Assume by way of contradiction that we are given an

	$\alpha$	$\beta$	$\gamma$
$\alpha$	1,0	0,1	1,0
$\beta$	0,1	1,0	1,0
$\gamma$	0,1	0,1	1,1

Figure 1: A non-generic two-player game

uncoupled, 1-recall, stationary strategy mapping that guarantees convergence to pure Nash equilibria when these exist. Note that at each of the nine action pairs, at least one of the two players is best-replying. Suppose the current state a(t) is such that player 1 is best-replying (the argument is symmetric for player 2). We claim that player 1 will play at t+1 the same action as in t (i.e., player 1 will not move). To see this consider a new game where the utility function of player 1 remains unaltered and the utility function of player 2 is changed in such a manner that the current state a(t) is the only pure Nash equilibrium of the new game. It is easy to check that in our game this can always be accomplished (for example, to make  $(\alpha, \gamma)$  the unique Nash equilibrium, change the payoff of player 2 in the  $(\alpha, \gamma)$  and  $(\gamma, \alpha)$  cells to, say, 2). The strategy mapping has 1-recall, so it must prescribe to the first player not to move in the new game (otherwise convergence to the unique pure equilibrium would be violated there). By uncoupledness, therefore, player 1 will not move in the original game either.

It follows that  $(\gamma, \gamma)$  can never be reached when starting from any other state: if neither player plays  $\gamma$  currently then only one player (the one who is

not best-replying) may move; if only one plays  $\gamma$  then the other player cannot move (since in all cases it is seen that he is best-replying). This contradicts our assumption.

The second example is the three-player game of Figure 2. There are three players i = 1, 2, 3, and each player has three actions  $\alpha, \beta, \gamma$ . Restricted to  $\alpha$ 

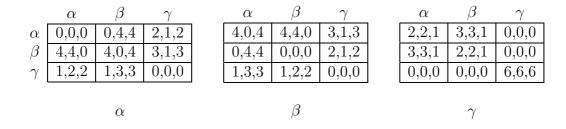


Figure 2: A generic three-player game

and  $\beta$  we essentially have the game of Jordan (1993) (see Hart and Mas-Colell (2003, Section III)), where every player i tries to mismatch the player i-1 (the predecessor of player 1 is player 3): he gets 0 if he matches and 4 if he mismatches. If all three players play  $\gamma$  then each one gets 6. If one player plays  $\gamma$  and the other two do not, the player that plays  $\gamma$  gets 1 and the other two get 3 each if they mismatch and 2 each if they match. If two players play  $\gamma$  and the third one does not then each one gets 0.

The only pure Nash equilibrium of this game is  $(\gamma, \gamma, \gamma)$ . Suppose that we start with all players playing  $\alpha$  or  $\beta$ , but not all the same; for instance,  $(\alpha, \beta, \alpha)$ . Then players 2 and 3 are best-replying, so only player 1 can move in the next period (this follows from uncoupledness as in the previous example). If he plays  $\alpha$  or  $\beta$  then we are in exactly the same position as before (with, possibly, the role of mover taken by player 2). If he moves to  $\gamma$  then the action configuration is  $(\gamma, \beta, \alpha)$ , at which both players 2 and 3 are best-replying and so, again, only player 1 can move. Whatever he plays next, we are back to situations already contemplated. In summary, every configuration that can be visited will only have at most one  $\gamma$ , and therefore the unique pure Nash equilibrium  $(\gamma, \gamma, \gamma)$  will never be reached.

**Remark.** In the three-player example of Figure 2, starting with  $(\alpha, \beta, \alpha)$ , the joint distribution of play cannot approach the distribution of a mixed Nash equilibrium, because neither  $(\alpha, \alpha, \alpha)$  nor  $(\beta, \beta, \beta)$  will ever be visited — but these action combinations have positive probability in every mixed Nash equilibrium (there are two such equilibria: in the first each player plays (1/2, 1/2, 0), and in the second each plays (1/4, 1/4, 1/2)).

As we noted above, the two-player example of Figure 1 is not generic. It turns out that in the case of only two players, *genericity* — in the sense that every player's best reply to pure actions is always unique — does help.

**Proposition 2** There exist uncoupled, 1-recall, stationary strategy mappings that guarantee almost sure convergence to pure Nash equilibria in every two-player generic game where such equilibria exist.

**Proof.** We define the strategy mapping of player 1 (and similarly for player 2) as follows. Let the state (i.e., the previous period's play) be  $a = (a^1, a^2) \in A = A^1 \times A^2$ . If  $a^1$  is a best reply to  $a^2$  according to  $u^1$ , then player 1 plays  $a^1$ ; otherwise, player 1 randomizes uniformly<sup>4</sup> over all his actions in  $A^1$ .

With these strategies, it is clear that each pure Nash equilibrium becomes an absorbing state of the resulting Markov chain.<sup>5</sup> Moreover, as we will show next, from any other state  $a=(a^1,a^2)\in A$  there is a positive probability to reach a pure Nash equilibrium in at most two steps. Indeed, since a is not a Nash equilibrium, then at least one of the players, say player 2, is not best-replying. Therefore there is a positive probability that in the next period player 2 plays  $\bar{a}^2$ , where  $(\bar{a}^1, \bar{a}^2)$  is a pure Nash equilibrium (which is assumed to exist), whereas player 1 plays again the same  $a^1$  of last period (this has probability one if player 1 was best-replying, and positive probability otherwise). The new state is thus  $(a^1, \bar{a}^2)$ . If  $a^1 = \bar{a}^1$ , we have reached the pure Nash equilibrium  $\bar{a}$ . If not, then player 1 is not best-replying — it is here that our genericity assumption is used: the unique best reply

<sup>&</sup>lt;sup>4</sup>I.e., the probability of each  $a^1 \in A^1$  is  $1/|A^1|$ . Of course, the uniform distribution may be replaced — here as well as in all other constructions in this paper — by any probability distribution with full support (i.e., such that every action has positive probability).

<sup>&</sup>lt;sup>5</sup>A standard reference for Markov chains is Feller (1968, Chapter XV).

to  $\bar{a}^2$  is  $\bar{a}^1$  — so now there is a positive probability that in the next period player 1 will play  $\bar{a}^1$  and player 2 will play  $\bar{a}^2$ , and thus, again, the pure Nash equilibrium  $\bar{a}$  is reached.

Therefore an absorbing state — i.e., a pure Nash equilibrium — must eventually be reached with probability one.  $\Box$ 

Interestingly, if we allow for longer recall the situation changes and we can present positive results for general games. In fact, for the case where pure Nash equilibria exist the contrast is quite dramatic, since allowing one more period of recall suffices.

**Theorem 3** There exist uncoupled, 2-recall, stationary strategy mappings that guarantee almost sure convergence to pure Nash equilibria in every game where such equilibria exist.

**Proof.** Let the state — i.e., the play of the previous two periods — be  $(a', a) \in A \times A$ . We define the strategy mapping of each player i as follows:

- if a' = a (i.e., if all players have played exactly the same actions in the past two periods) and  $a^i$  is a best reply of player i to  $a^{-i}$  according to  $u^i$ , then player i plays  $a^i$  (i.e., he plays again the same action);
- in all other cases, player i randomizes uniformly over  $A^i$ .

To prove our result, we partition the state space  $S = A \times A$  of the resulting Markov chain into four regions:

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\begin{array}{lll} S_1 &:=& \{(a,a) \in A \times A : a \text{ is a Nash equilibrium}\};\\ S_2 &:=& \{(a',a) \in A \times A : a' \neq a \text{ and } a \text{ is a Nash equilibrium}\};\\ S_3 &:=& \{(a',a) \in A \times A : a' \neq a \text{ and } a \text{ is not a Nash equilibrium}\};\\ S_4 &:=& \{(a,a) \in A \times A : a \text{ is not a Nash equilibrium}\}. \end{array}
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Clearly, each state in  $S_1$  is absorbing. Next, we claim that all other states are transient: there is a positive probability to reach a state in  $S_1$  in finitely many periods. Indeed:

- At each state (a', a) in  $S_2$  all players randomize; hence there is positive probability that next period they will play a and so the next state will be (a, a), which belongs to  $S_1$ .
- At each state (a', a) in  $S_3$  all players randomize; hence there is positive probability that next period they will play a pure Nash equilibrium  $\bar{a}$  (which exists by assumption) and so the next state will be  $(a, \bar{a})$ , which belongs to  $S_2$ .
- At each state (a, a) in  $S_4$  at least one player is not best-replying and thus randomizing; hence there is positive probability that the next period play will be some  $a' \neq a$  and so the next state will be (a, a'), which belongs to  $S_2 \cup S_3$ .

In all cases there is thus positive probability of reaching an absorbing state in  $S_1$  in at most three steps. Once such a state (a, a), where a is a pure Nash equilibrium, is reached (this happens eventually with probability one), the players will continue to play a every period.

Thus extremely simple strategies may nevertheless guarantee convergence to pure Nash equilibria. The strategies defined above may be viewed as a combination of search and testing. The search is a standard random search; the testing is done individually, but in a coordinated manner: the players wait until a certain "pattern" (a repetition) is observed, at which point each one applies a "rational" test (he checks whether or not he is best-replying). Finally, the pattern is self-replicating once the desired goal (a Nash equilibrium) is reached. (This structure will appear also, in a slightly more complex way, in the case of mixed equilibria; see the proofs of Proposition 4 and Theorem 5 below.)

# 4 Mixed Equilibria

We come next to the general case (where only the existence of mixed Nash equilibria is guaranteed). The convergence will now be to approximate equilibria. To this effect, assume that there is a bound M on payoffs; i.e., the

payoff functions all satisfy  $|u^i(a)| \leq M$  for all action combinations  $a \in A$  and all players i.

Given a history of play, we will denote by  $\Phi_t$  the empirical frequency distribution in the first t periods:  $\Phi_t[a] := |\{1 \leq \tau \leq t : a(\tau) = a\}|/t$  for each  $a \in A$ , and similarly  $\Phi_t[a^i] := |\{1 \leq \tau \leq t : a^i(\tau) = a^i\}|/t$  for each i and  $a^i \in A^i$ . We will refer to  $(\Phi_t[a])_{a \in A} \in \Delta(A)$  as the joint distribution of play, i and to  $(\Phi_t[a^i])_{a^i \in A^i} \in \Delta(A^i)$  as the marginal distribution of play of player i (up to time i).

**Proposition 4** For every M and  $\varepsilon > 0$  there exists an integer R and an uncoupled, R-recall, stationary strategy mapping that guarantees, in every game with payoffs bounded by M, almost sure convergence of the marginal distributions of play to Nash  $\varepsilon$ -equilibria; i.e., for almost every history of play there exists a Nash  $\varepsilon$ -equilibrium of the stage game  $x = (x^1, x^2, ..., x^N)$  such that, for every player i and every action  $a^i \in A^i$ ,

$$\lim_{t \to \infty} \Phi_t[a^i] = x^i(a^i). \tag{1}$$

Of course, different histories may lead to different  $\varepsilon$ -equilibria. The length of the recall R depends on the precision  $\varepsilon$  and the bound on payoffs M (as well as on number of players N and the number of actions  $|A^i|$ ).

**Proof.** Given  $\varepsilon > 0$ , let K be such that<sup>7</sup>

$$\left[ ||x^i - y^i|| \le \frac{1}{K} \text{ for all } i \right] \implies \left[ |u^i(x) - u^i(y)| \le \varepsilon \text{ for all } i \right]$$
 (2)

for  $x^i, y^i \in \Delta(A^i)$  and  $|u^i(a)| \leq M$  for all  $a \in A$ . Let  $\bar{y} = (\bar{y}^1, \bar{y}^2, ..., \bar{y}^N)$  be a Nash  $2\varepsilon$ -equilibrium, such that all probabilities are multiples of 1/K (i.e.,  $K\bar{y}^i(a^i)$  is an integer for all  $a^i$  and all i). Such a  $\bar{y}$  always exists: take a 1/K-approximation of a Nash equilibrium and use (2). Given such a Nash  $2\varepsilon$ -equilibrium  $\bar{y}$ , let  $(\bar{a}_1, \bar{a}_2, ..., \bar{a}_K) \in A \times A \times ... \times A$  be a fixed sequence of action

<sup>&</sup>lt;sup>6</sup>Also known as the "empirical (or sample) distribution of play."

<sup>&</sup>lt;sup>7</sup>We use the maximum  $(\ell^{\infty})$  norm on  $\Delta(A^i)$ , i.e.,  $||x^i - y^i|| := \max_{a^i \in A^i} |x^i(a^i) - y^i(a^i)|$ ; it is easy to check that  $K \geq M \sum_i |A^i|/\varepsilon$  suffices for (2).

combinations of length K whose marginals are precisely  $\bar{y}^i$  (i.e., each action  $a^i$  of each player i appears  $K\bar{y}^i(a^i)$  times in the sequence  $(\bar{a}_1^i, \bar{a}_2^i, ..., \bar{a}_K^i)$ ).

Take R=2K. The construction parallels the one in the Proof of Theorem 3. A state is a history of play of length 2K, i.e.,  $s=(a_1,a_2,...,a_{2K})$  with  $a_k \in A$  for all k. The state s is K-periodic if  $a_{K+k}=a_k$  for all k=1,2,...,K. Given s, for each player i we denote by  $z^i \in \Delta(A^i)$  the frequency distribution of the last K actions of i, i.e.,  $z^i(a^i) := |\{K+1 \le k \le 2K : a_k^i = a^i\}/K$  for each  $a^i \in A^i$ ; put  $z=(z^1,z^2,...,z^N)$ .

We define the strategy mapping of each player i as follows:

- if the current state s is K-periodic and  $z^i$  is a  $2\varepsilon$ -best reply to  $z^{-i}$ , then player i plays  $a_1^i = a_{K+1}^i$  (i.e., continues his K-periodic play);
- in all other cases player i randomizes uniformly over  $A^i$ .

Partition the state space S consisting of all sequences over A of length 2K into four regions:

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S_1 := \{s \text{ is } K\text{-periodic and } z \text{ is a Nash } 2\varepsilon\text{-equilibrium}\};
S_2 := \{s \text{ is not } K\text{-periodic and } z \text{ is a Nash } 2\varepsilon\text{-equilibrium}\};
S_3 := \{s \text{ is not } K\text{-periodic and } z \text{ is not a Nash } 2\varepsilon\text{-equilibrium}\};
S_4 := \{s \text{ is } K\text{-periodic and } z \text{ is not a Nash } 2\varepsilon\text{-equilibrium}\}.
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We claim that the states in  $S_1$  are persistent and K-periodic, and all other states are transient. Indeed, once a state s in  $S_1$  is reached, the play moves in a deterministic way through the K cyclic permutations of s, all of which have the same z — and so, for each player i, his marginal distribution of play will converge to  $z^i$ . At a state s in  $S_2$  every player randomizes, so there is positive probability that everyone will play K-periodically, leading in  $r = \max\{1 \le k \le K : a_{K+k} \ne a_k\}$  steps to  $S_1$ . At a state s in  $S_3$ , there is positive probability of reaching  $S_2$  in K+1 steps: in the first step the play is some  $a \ne a_{K+1}$ , and, in the next K steps, a sequence  $(\bar{a}_1, \bar{a}_2, ..., \bar{a}_K)$  corresponding to a Nash  $2\varepsilon$ -equilibrium. Finally, from a state in  $S_4$  there is positive probability of moving to a state in  $S_2 \cup S_3$  in one step.

Proposition 4 is not entirely satisfactory, because it does not imply that the *joint* distributions of play converge to joint distributions induced by Nash approximate equilibria. For this to happen, the joint distribution needs to be (in the limit) the product of the marginal distributions (i.e., independence among the players' play is required). But this is not the case in the construction in the Proof of Proposition 4 above, where the players' actions become "synchronized" — rather than independent — once an absorbing cycle is reached. A more refined proof is thus needed to obtain the stronger conclusion of the following theorem on the convergence of the joint distributions.

**Theorem 5** For every M and  $\varepsilon > 0$  there exists an integer R and an uncoupled, R-recall, stationary strategy mapping that guarantees, in every game with payoffs bounded by M, the almost sure convergence of the joint distributions of play to Nash  $\varepsilon$ -equilibria; i.e., for almost every history of play there exists a Nash  $\varepsilon$ -equilibrium of the stage game  $x = (x^1, x^2, ..., x^N)$  such that, for every action combination  $a = (a^1, a^2, ..., a^N) \in A$ ,

$$\lim_{t \to \infty} \Phi_t[a] = \prod_{i=1}^N x^i(a^i). \tag{3}$$

Moreover, the t-period occurrence probabilities Pr[a(t) = a] also converge to the same Nash  $\varepsilon$ -equilibrium x:

$$\lim_{t \to \infty} \Pr[a(t) = a] = \prod_{i=1}^{N} x^{i}(a^{i})$$
(4)

for every  $a \in A$ .

Since the proof of Theorem 5 is relatively intricate, we relegate it to the Appendix. Of course, (1) follows from (3). Note that neither (4) nor its marginal implications,

$$\lim_{t \to \infty} \Pr[a^i(t) = a^i] = x^i(a^i) \tag{5}$$

for all i, hold for the construction of Proposition 4 (again, due to periodicity).

Now (5) says that the overall (unconditional) probabilities of play converge almost surely to Nash  $\varepsilon$ -equilibria. It does not say the same, however, about the (conditional-on-history) behavior probabilities  $\Pr[a^i(t) = a^i \mid a(1), ..., a(t-1)] = f^i(a(1), ..., a(t-1))(a^i)$ . We next show that this cannot be guaranteed in general.

**Theorem 6** For every small enough<sup>8</sup>  $\varepsilon > 0$ , there are no uncoupled, finite recall, stationary strategy mappings that guarantee, in every game, the almost sure convergence of the behavior probabilities to Nash  $\varepsilon$ -equilibria.

**Proof.** Choose a stage game U with a unique, completely mixed Nash equilibrium, and assume that a certain pure action combination, call it  $\bar{a} \in A$ , is such that  $\bar{a}^1$  is the unique best reply of player 1 to  $\bar{a}^{-1} = (\bar{a}^2, ..., \bar{a}^N)$ . Let U' be another game where the payoff function of player 1 is the same as in U, and the payoff function of every other player  $i \neq 1$  depends only on  $a^i$  and has a unique global maximum at  $\bar{a}^i$ . Then  $\bar{a}$  is the unique Nash equilibrium of U'. Take  $\varepsilon > 0$  small enough so that all Nash  $\varepsilon$ -equilibria of U are completely mixed, and moreover there exists  $\rho > 0$  such that, for any two Nash  $\varepsilon$ -equilibria x and y of U and U', respectively, we have  $0 < x^i(\bar{a}^i) < \rho < y^i(\bar{a}^i)$  for all players i (recall that  $x^i(\bar{a}^i) < 1 = y^i(\bar{a}^i)$  when x and y are the unique Nash equilibria of U and U', respectively).

We argue by contradiction and assume that for some R there is an uncoupled, R-recall, stationary strategy mapping f for which the stated convergence does in fact obtain.

Consider now the history  $\bar{\mathbf{a}} = (\bar{a}, \bar{a}, ..., \bar{a})$  of length R that consists of R repetitions of  $\bar{a}$ . The behavior probabilities have been assumed to converge (a.s.) to Nash  $\varepsilon$ -equilibria, which, in both games, always give positive probability to the actions  $\bar{a}^i$ . Hence the state  $\bar{\mathbf{a}}$  has a positive probability of occurring after any large enough time T. Therefore, in particular at this state  $\bar{\mathbf{a}}$ , the behavior probabilites must be close to Nash  $\varepsilon$ -equilibria. Now all Nash  $\varepsilon$ -equilibria x of U satisfy  $x^1(\bar{a}^1) < \rho$ , so the behavior probability of player 1 at state  $\bar{\mathbf{a}}$  must also satisfy this inequality, i.e.,  $f^1(\bar{\mathbf{a}}; u^1)(\bar{a}^1) < \rho$ . But the

<sup>&</sup>lt;sup>8</sup>I.e., for all  $\varepsilon < \varepsilon_0$  (where  $\varepsilon_0$  may depend on N and  $(|A^i|)_{i=1}^N$ ).

same argument applied to U' (where player 1 has the same payoff function  $u^1$  as in U) implies  $f^1(\bar{\mathbf{a}}; u^1)(\bar{a}^1) > \rho$  (since this inequality is satisfied by all Nash  $\varepsilon$ -equilibria of U'). This contradiction proves our claim.

## 5 Discussion and Comments

This section includes some further comments, particularly on the relevant literature.

- (a) Convergence: We emphasize that throughout this paper we have sought a very strong form of convergence, namely, almost sure convergence to a point.<sup>9</sup> One could consider seeking weaker forms of convergence (as has been done in the related literature): almost sure convergence to the convex hull of the set of Nash  $\varepsilon$ -equilibria, or convergence in probability, or " $1-\varepsilon$  of the time being an  $\varepsilon$ -equilibrium," and so on. Conceivably, the use of weaker forms of convergence may have a theoretical payoff in other aspects of the analysis. Comments on some of these trade-offs are included in the next two remarks.
- (b) Behavior probabilities: Theorem 6 has established that we cannot hope to obtain in all generality the almost sure convergence of the actual play (or behavior) probabilities.<sup>10</sup> Can this be obtained for some of the weaker notions of Remark (a), and if so, then at what cost (in terms, for example, of the type of convergence obtained for the sample distribution of play)? The construction of Theorem 5 does not yield good results in this respect: the joint distributions of play, and also the occurrence probabilities, converge in a strong sense, but there is no convergence, even "in probability," of the behavior probabilities. The results of Foster and Young (2003a) are of direct relevance to this point.
- (c)  $Unknown\ game$ : Suppose that the players observe, not the history of play, but only their own realized payoffs; i.e., for each player i and time t

<sup>&</sup>lt;sup>9</sup>The negative results of Theorems 1 and 6 also hold for certain weaker forms of convergence.

 $<sup>^{10}{\</sup>rm This}$  convergence is obtained for pure equilibria — see the constructions of Proposition 2 and Theorem 3.

the strategy is  $f_t^i(u^i(a(1)), u^i(a(2)), ..., u^i(a(t-1))$  (in fact, the player may know nothing about the game being played but his set of actions). What results can be obtained in this case? It appears that, for any positive result, experimentation even at (apparent) Nash equilibria will be indispensable. This suggests, in particular, that the best sort of convergence to hope for is some kind of convergence in probability as mentioned in Remark (a). On this point see Foster and Young (2003b).

(d) Foster and Young: The current paper is not the first one where, within the span of what we call uncoupled dynamics, stochastic moves and the possibility of recalling the past have been brought to bear on the formulation of dynamics leading to Nash equilibria. The pioneers were Foster and Young (2003a), followed by Foster and Young (2003b), Kakade and Foster (2003), and Germano and Lugosi (2004).

The motivation of Foster and Young and our motivation are not entirely the same. They want to push to its limits the "learning with experimentation" paradigm (which does not allow direct exhaustive search procedures that, in our terminology, are not of an uncoupled nature). We start from the uncoupledness property and try to demarcate the border between what can and what cannot be done with such dynamics.

In terms of results, some of the differences between Foster and Young's work and ours have already been mentioned in previous remarks: we use a stronger notion of convergence, but do not obtain the convergence of the behavior strategies, nor do we handle the "unknown game" situation of Remark (c).

- (e) Finiteness: In this paper we "make the past finite" by means of a finite recall assumption. This is standard but it is not the only way to do so. One could allow for a finite number of updatable statistics, or for carrying a finite number of "instructions" from the past, or for finite automata. It would be of interest to determine to what extent our conclusions generalize to these environments.
- (f) Adaptive strategies: Suppose we were to require in addition that the strategies of the players be "adaptive" in one way or another. For example, at time t player i could randomize only over actions that improve i's payoff given

some sort of "expected" behavior of the other players at t, or over actions that would have yielded a better payoff if played at t-1, or if played every time in the past that the action at t-1 was played, or if played every time in the past (these last two are in the style of "regret-based" strategies; see Hart (2004) for a survey). Would the positive results of this paper still obtain? Note that such adaptive or monotonicity-like conditions severely restrict the possibilities of "free experimentation" that drive the positive results obtained here.

(g) Correlated equilibria: We know that there are uncoupled strategy mappings with the property that the joint distributions of play converge almost surely to the set of correlated equilibria (see Foster and Vohra (1997), Hart and Mas-Colell (2000), Hart (2004), and the book of Young (2004)). Strictly speaking, those strategies do not have finite recall, but enjoy a closely related property: they depend (in a stationary way) on a finite number of summary, and easily updatable, statistics from the past; see Remark (e). The results of these papers differ from those of the current paper in several respects. First, the convergence there is to a set, whereas here it is to a point. Second, the convergence there is to correlated equilibria, whereas here it is to Nash equilibria. And third, the strategies there are natural, adaptive, heuristic strategies, while in this paper we are dealing with forms of exhaustive search. An issue for further study is to what extent the contrast could be captured by an analysis of the speeds of convergence.

# Appendix: Proof of Theorem 5

As pointed out in Section 4, the problem with our construction in the Proof of Proposition 4 is that it leads to periodic and synchronized behavior. To avoid this we introduce small perturbations, independently for each player: once in a while, there is a positive probability of repeating the previous period's action rather than continuing the periodic play.<sup>11</sup> To guarantee that this does not eventually change the frequencies of play, we now use three

<sup>&</sup>lt;sup>11</sup>This kind of perturbation was suggested by Benji Weiss.

repetitions rather than two and make sure that the basic periodic play can always be recognized from the R-history (since our players cannot use any additional "notes" or "instructions"  $^{12}$ ).

**Proof of Theorem 5**. We take R = 3K, where K > 2 is chosen so as to satisfy (2). Consider sequences  $\mathbf{b} = (b_1, b_2, ..., b_{3K})$  of length 3K over an arbitrary finite set B (i.e.,  $b_k \in B$  for all k). We distinguish two types of such sequences:

- **Type E** ("Exact"): The sequence is K-periodic, i.e.,  $b_{K+k} = b_k$  for all  $1 \le k \le 2K$ . Thus **b** consists of three repetitions of the basic K-sequence  $\mathbf{c} := (b_1, b_2, ..., b_K)$ .
- **Type D** ("Delay"): The sequence is not of type E, and there is  $2 \le d \le 3K$  such that  $b_d = b_{d-1}$  and if we drop the element  $b_d$  from the sequence  $\mathbf{b}$  then the remaining sequence  $\mathbf{b}_{-d} = (b_1, ..., b_{d-1}, b_{d+1}, ..., b_{3K})$  of length 3K-1 is K-periodic. Again, let  $\mathbf{c}$  denote the basic K-sequence,  $^{13}$  so  $\mathbf{b}_{-d}$  consists of three repetitions of  $\mathbf{c}$ , except for the missing last element. Think of  $b_d$  as a "delay" element.

We claim that the basic sequence of a sequence  $\mathbf{b}$  of type D is uniquely defined. Indeed, assume that  $\mathbf{b}_{-d}$  and  $\mathbf{b}_{-d'}$  are both K-periodic, with corresponding basic sequences  $\mathbf{c} = (c_1, c_2, ..., c_K)$  and  $\mathbf{c}' = (c'_1, c'_2, ..., c'_K)$ , and d < d'. If  $d \ge K + 1$ , then the first K coordinates of  $\mathbf{b}$  determine the basic sequence:  $(b_1, b_2, ..., b_K) = \mathbf{c} = \mathbf{c}'$ . If  $d' \le 2K$ , then the last K coordinates determine this:  $(b_{2K+1}, b_{2K+2}, ..., b_{3K}) = (c_K, c_1, ..., c_{K-1}) = (c'_K, c'_1, ..., c'_{K-1})$ , so again  $\mathbf{c} = \mathbf{c}'$ . If neither of these two hold, then  $d \le K$  and  $d' \ge 2K + 1$ . Without loss of generality assume that we took d' to be maximal such that  $\mathbf{b}_{-d'}$  is K-periodic, and let d' = 2K + r (where  $1 \le r \le K$ ). Now  $b_{d'-1} = c'_{r-1}$  and  $b_{d'} = c_{r-1}$ , so  $c_{r-1} = c'_{r-1}$  (since  $b_{d'} = b_{d'-1}$ ). But  $c_{r-1} = b_{K+r} = c'_r$  (since d < K + r < d'; if r = 1 put  $r - 1 \equiv K$ ), so  $c'_{r-1} = c'_r$ . If d' < 3K, this last equality implies that  $b_{d'} = b_{d'+1}$ , so  $\mathbf{b}_{-(d'+1)}$  is also K-periodic (with the

 $<sup>^{12}\</sup>mathrm{As}$  would be the case, for instance, were the strategies of finite memory; see Remark (e) in Section 5.

 $<sup>^{13}</sup>$ As we will see immediately below, **c** is well-defined (even though d need not be unique).

same basic sequence  $\mathbf{c}'$ ), contradicting the maximality of d'. If d' = 3K, the equality becomes  $c'_{K-1} = c'_{K}$ , which implies that the sequence  $\mathbf{b}$  is in fact of type E (it consists of three repetitions of  $\mathbf{c}'$ ), again a contradiction.

Given a sequence **b** of type E or D, the frequency distribution of its basic K-sequence **c**, i.e.,  $w \in \Delta(B)$  where  $w(b) := |\{1 \le k \le K : c_k = b\}|/K$  for each  $b \in B$ , will be called the basic frequency distribution of **b**.

To define the strategies, let  $\mathbf{a} = (a_1, a_2, ..., a_{3K}) \in A \times A \times ... \times A$  be the state — a history of action combinations of length 3K — and put  $\mathbf{a}^i := (a_1^i, a_2^i, ..., a_{3K}^i)$  for the corresponding sequence of actions of player i. When  $\mathbf{a}^i$  is of type E or D, we denote by  $y^i \in \Delta(A^i)$  its basic frequency distribution. If for each player i the sequence  $\mathbf{a}^i$  is of type E or D, we will say that the state  $\mathbf{a}$  is regular; otherwise we will call  $\mathbf{a}$  irregular.

The strategy of player i is defined as follows.

- (\*) If the state **a** is regular, the basic frequency  $y^i$  is a  $4\varepsilon$ -best reply to the basic frequencies of the other players  $y^{-i} = (y^j)_{j\neq i}$ , and the sequence  $\mathbf{a}^i$  is of type E, then with probability 1/2 play  $a_1^i$  (i.e., continue the K-periodic play), and with probability 1/2 play  $a_{3K}^i$  (i.e., introduce a "delay" period by repeating the previous period's action).
- (\*\*) If the state **a** is regular, the basic frequency  $y^i$  is a  $4\varepsilon$ -best reply to the basic frequencies of the other players  $y^{-i} = (y^j)_{j \neq i}$ , and the sequence  $\mathbf{a}^i$  is of type D, then play the last element of the basic sequence  $\mathbf{c}$ , which is  $a_K^i$  if d > K and  $a_{K+1}^i$  if  $d \leq K$  (i.e., continue the K-periodic play).
- (\*\*\*) In all other cases randomize uniformly over  $A^i$ .

As in the Proof of Proposition 4, given a state **a**, for each player i let  $z^i \in \Delta(A^i)$  denote the frequency distribution of  $(a^i_{2K+1}, a^i_{2K+2}, ..., a^i_{3K})$ , the last K actions of i, and put  $z = (z^1, z^2, ..., z^N)$ ; also,  $y = (y^1, y^2, ..., y^N)$  is the N-tuple of the basic frequency distributions. We partition the state space S,

 $<sup>^{14}</sup>$ Some players' sequences may be E, and others', D (moreover, they may have different d's); therefore **a** itself need not be E or D.

which consists of all sequences **a** of length 3K over A, into four regions:

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S_1 := \{ \mathbf{a} \text{ is regular and } y \text{ is a Nash } 4\varepsilon\text{-equilibrium} \};
S_2 := \{ \mathbf{a} \text{ is irregular and } z \text{ is a Nash } 2\varepsilon\text{-equilibrium} \};
S_3 := \{ \mathbf{a} \text{ is regular and } y \text{ is not a Nash } 4\varepsilon\text{-equilibrium} \};
S_4 := \{ \mathbf{a} \text{ is irregular and } z \text{ is not a Nash } 2\varepsilon\text{-equilibrium} \}.
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We analyze each region in turn.

Claim 1. All states in  $S_1$  are ergodic.<sup>15</sup>

**Proof.** Let  $\mathbf{a} \in S_1$ . For each player i, let  $\mathbf{c}^i = (c_1^i, c_2^i, ..., c_K^i)$  be the basic sequence of  $\mathbf{a}^i$ , with  $y^i$  the corresponding basic frequency distribution. The strategies are such that the sequence of i in the next period is also of type E or D (by (\*) and (\*\*)), with basic sequence that is the cyclical permutation of  $\mathbf{c}^i$  by one step  $(c_2,...,c_K,c_1)$ , except when  $\mathbf{a}^i$  is of type D with d=2, in which case it remains unchanged. Therefore the basic frequency distribution  $y^i$  does not change, and the new state is also in  $S_1$ . Hence  $S_1$  is a closed set, and once it is reached the conditions of regularity and  $4\varepsilon$ -best-replying for each player will always continue to be automatically satisfied; thus each player's play in  $S_1$  depends only on whether his own sequence is of type E or D (again, see (\*) and (\*\*)). Therefore in the region  $S_1$  the play becomes independent among the players, and the Markov chain restricted to  $S_1$  is the product of N independent Markov chains, one for each player. Specifically, the state space  $S_1^i$  of the Markov chain of player i consists of all sequences of length 3K over  $A^i$  that are of type E or D, and the transition probabilities are defined as in (\*) and (\*\*), according to whether the sequence is of type E or D, respectively.

We thus analyze each i separately. Let  $\mathbf{a}^i$  be a 3K-sequence in  $S_1^i$  with basic sequence  $\mathbf{c}^i$ . The closure of  $\mathbf{a}^i$  (i.e., the minimal closed set containing  $\mathbf{a}^i$ ) consists of all  $\tilde{\mathbf{a}}^i$  in  $S_1^i$  whose basic sequence is one of the K cyclical permutations of  $\mathbf{c}^i$ : any such state can be reached from any other in finitely

 $<sup>^{15}</sup>$ I.e., aperiodic and persistent; see Feller (1968, Sections XV.4-6) for these and the other Markov chains' concepts that we use below.

many steps (for instance, it takes at most 3K-1 steps to get to a sequence of type E, then at most K-1 steps to the appropriate cyclical permutation, and then another 3K steps to introduce a delay and wait until it reaches the desired place in the sequence).

Next, the states in  $S_1^i$  are aperiodic. Indeed, if the basic sequence  $\mathbf{c}^i$ is constant (i.e.,  $\mathbf{c}^i = (\hat{a}^i, \hat{a}^i, ..., \hat{a}^i)$  for some  $\hat{a}^i \in A^i$ ), then the constant sequence  $(\hat{a}^i, \hat{a}^i, ..., \hat{a}^i)$  of length 3K is an absorbing state (since the next play of i will always be  $\hat{a}^i$  by (\*)), and thus aperiodic. If  $\mathbf{c}^i$  is not constant, then assume without loss of generality that  $c_1^i \neq c_K^i$  (if not, take an appropriate cyclical permutation of  $\mathbf{c}^i$ , which keeps us in the same minimal closed set). Let  $\mathbf{a}^i$  be the sequence of type E that consists of three repetitions of  $\mathbf{c}^i$ . Starting at  $\mathbf{a}^i$ , there is a positive probability that  $\mathbf{a}^i$  is reached again in K steps, by always making the first choice in (\*) (i.e., playing K-periodically, with no delays). However, there is also a positive probability of returning to  $\mathbf{a}^i$  in 3K+1 steps, by always making the first choice in (\*), except for the initial choice which introduces a delay (after 3K additional steps the delay coordinate is no longer part of the state and we return to the original sequence  ${}^{16}$   $\mathbf{a}^{i}$ ). But K and 3K+1 are relatively prime, so the state  $\mathbf{a}^{i}$  is aperiodic. Therefore, every minimal closed set contains an aperiodic state, and all states are aperiodic.

Returning to the original Markov chain (over N-tuples of actions), the product of what we have shown is that, to each combination of basic sequences  $(\mathbf{c}^1, \mathbf{c}^2, ..., \mathbf{c}^N)$  whose frequency distributions constitute a Nash  $4\varepsilon$ -equilibrium, there corresponds an ergodic set<sup>17</sup> consisting of all states with basic sequences that are, for each i, some cyclic permutation of  $\mathbf{c}^i$ . The set  $S_1$  is precisely the union of all these ergodic sets.

The next three claims show that all states outside  $S_1$  are transient: from any such state there is a positive probability of reaching  $S_1$  in finitely many steps.

<sup>&</sup>lt;sup>16</sup>The condition  $c_1^i \neq c_K^i$  is needed in order for the delay action,  $c_K^i$ , to be different from the K-periodic action,  $c_L^i$ .

<sup>&</sup>lt;sup>17</sup>I.e., a minimal closed and aperiodic set.

Claim 2. Starting from any state in  $S_2$ , there is a positive probability that a state in  $S_1$  is reached in at most 2K steps.

**Proof.** Let  $\mathbf{a} \in S_2$ . Since at state  $\mathbf{a}$  case (\*\*\*) applies to every player i, there is a positive probability that i will play  $a_{2K+1}^i$  (i.e., play K-periodically). If the new state  $\mathbf{a}'$  is regular, then for each i the frequency distribution  $y^i$  of the resulting basic sequence either equals the frequency distribution  $z^i$  of the last K periods, or differs from it by 1/K (in the maximum norm). But  $z^i$  is a  $2\varepsilon$ -best reply to  $z^{-i}$ , which implies that  $y^i$  is a  $4\varepsilon$ -best reply to  $y^{-i}$  by (2) — and so  $\mathbf{a}' \in S_1$ . If the new state  $\mathbf{a}'$  is irregular, then  $\mathbf{a}' \in S_2$ . Again, at  $\mathbf{a}'$  there is a positive probability that every player will play K-periodically. Continuing this way, we must reach at some point a regular state — since after 2K such steps the sequence of each player is surely of type E — a state that is therefore in  $S_1$ .

Claim 3. Starting from any state in  $S_3$ , there is a positive probability that a state in  $S_2$  is reached in at most 5K + 1 steps.

**Proof.** Let  $\mathbf{a} \in S_3$ . There is a positive probability that every player will continue to play his basic sequence K-periodically, with no delays (this has probability 1/2 in (\*), 1 in (\*\*), and  $1/|A^i$  in (\*\*\*)). After at most 3K+1 steps, we get a sequence of type E for every player (since all the original delay actions are no longer part of the state). During these steps the basic frequencies  $y^i$  did not change, so there is still one player, say player 1, such that  $y^1$  is not a  $4\varepsilon$ -best reply to  $y^{-1}$ . So case (\*\*\*) applies to player 1, and thus there is a positive probability that he next plays an action  $\hat{a}^1$  that satisfies  $1 - y^1(\hat{a}^1) > 1/K$  (for instance, let  $\hat{a}^1 \in A^1$  have minimal frequency in  $y^1$ , then  $y^1(\hat{a}^1) \leq 1/|A^1| \leq 1/2$  and so  $1 - y^1(\hat{a}^1) \geq 1/2 > 1/K$ ). The sequence of every other player  $i \neq 1$  is of type E, so with positive probability i plays  $a_{2K+1}^i$  (this has probability 1/2 if (\*) and  $1/|A^i|$  if (\*\*\*)), and thus the sequence of i remains of type E. With positive probability this continues for K periods for all players  $i \neq 1$ . As for player 1, note that  $y^{-1}$  does not

 $<sup>^{18} \</sup>text{This}$  is where K>2 is used. Note that  $|A^1|\geq 2,$  since otherwise player 1 would always be best-replying.

change (since all other players  $i \neq 1$  play K-periodically and so their  $y^i$  does not change). If at any point during these K steps the sequence of player 1 turns out to be of type E or D, then it contains at least two repetitions of the original basic sequence (recall that we started with three repetitions, and we have made at most K steps), so the basic frequency is still  $y^1$ ; but  $y^1$  is not a  $4\varepsilon$ -best reply to  $y^{-1}$ , so we are in case (\*\*\*). If the sequence is not of type E or D then of course we are in case (\*\*\*) — so case (\*\*\*) always applies during these K steps, and there is a positive probability that player 1 will always play  $\hat{a}^1$ .

But after K periods the sequence of player 1 is for sure neither of type E nor D: the frequency of  $\hat{a}^1$  in the last K periods equals 1, and in the middle K periods it equals  $y^1(\hat{a}^1)$ , and these differ by more than 1/K (whereas in a sequence of type E or D, any two blocks of length K may differ in frequencies by at most 1/K). Now these two K-blocks remain part of the state for K more periods, during which the sequence of player 1 can thus be neither E nor D, and so the state is irregular. Hence case (\*\*\*) applies to all players during these K periods, and there is a positive probability that each player i plays a K-sequence whose frequency is  $\bar{y}^i$ , where  $(\bar{y}^1, \bar{y}^2, ..., \bar{y}^N)$  is a Nash  $2\varepsilon$ -equilibrium (see the beginning of the Proof of Proposition 4). So, finally, after at most (3K+1)+K+K steps, a state in  $S_2$  is reached.

Claim 4. Starting from any state in  $S_4$ , there is a positive probability that a state in  $S_1 \cup S_2 \cup S_3$  is reached in at most K steps.

**Proof.** At  $\mathbf{a} \in S_4$ , case (\*\*\*) applies to every player. There is therefore a positive probability that each player i plays according to a K-sequence with frequency  $\bar{y}^i$ , where again  $(\bar{y}^1, \bar{y}^2, ..., \bar{y}^N)$  is a fixed Nash  $2\varepsilon$ -equilibrium. Continue this way until either a regular state (in  $S_1$  or  $S_3$ ) is reached, or, if not, then after at most K steps the state is in  $S_2$ .

Combining the four claims implies that almost surely one of the ergodic sets, all of which are subsets of  $S_1$ , will eventually be reached. It remains to prove (3) and (4).

Let  $Q \subset S_1$  be an ergodic set. As we saw in the Proof of Claim 1, all

states in Q have the same basic frequency distributions  $(y^1, y^2, ..., y^N)$ . The independence among players in  $S_1$  implies that  $Q = Q^1 \times Q^2 \times ... \times Q^N$ , where  $Q^i \subset S_1^i$  is an ergodic set for the Markov chain of player i. For each  $\mathbf{a}^i \in Q^i$ , let  $x_{\mathbf{a}^i}^i \in \Delta(A^i)$  be the frequency distribution of all 3K coordinates of  $\mathbf{a}^i$ , then  $||x_{\mathbf{a}^i}^i - y^i|| \leq 1/(3K)$  (they may differ when the sequence contains a delay). Let  $\mu^i$  be the unique invariant probability measure on  $Q^i$ , then the average frequency distribution  $x^i := \sum_{\mathbf{a}^i \in Q^i} \mu^i(\mathbf{a}^i) x_{\mathbf{a}^i}^i \in \Delta(A^i)$  also satisfies  $||x^i - y^i|| \leq 1/(3K)$ . But  $(y^1, y^2, ..., y^N)$  is a Nash  $4\varepsilon$ -equilibrium, and so  $(x^1, x^2, ..., x^N)$  is a Nash  $6\varepsilon$ -equilibrium (by (2)).

Once the ergodic set  $Q^i$  is reached, the probability of occurrence of each state  $\mathbf{a}^i = (a^i_1, a^i_2, ..., a^i_{3K})$  in  $Q^i$  converges to its invariant probability (see Feller (1968, Section XV.7)):

$$\lim_{t \to \infty} \Pr[a^i(t+k) = a^i_k \text{ for } k = 1, 2, ..., 3K] = \mu^i(\mathbf{a}^i).$$

Projecting on the k-th coordinate yields, for every  $a^i \in A^i$ ,

$$\lim_{t \to \infty} \Pr[a^i(t) = a^i] = \sum_{\mathbf{a}^i \in Q^i : a_k^i = a^i} \mu^i(\mathbf{a}^i),$$

so, in particular, the limit on the left-hand side exists. Averaging over k = 1, 2, ..., 3K yields on the right-hand side  $\sum_{\mathbf{a}^i \in Q^i} \mu^i(\mathbf{a}^i) \, x^i_{\mathbf{a}^i}(a^i)$ , which equals  $x^i(a^i)$ ; this proves (5), from which (4) follows by independence.

A similar argument applies to the limit of the long-run frequencies  $\Phi_t$ . This completes the proof of Theorem 5.

## References

Feller, W. (1968), An Introduction to Probability and its Applications, Volume I, Third Edition, New York: J. Wiley.

Foster, D. P. and R. V. Vohra (1997), "Calibrated Learning and Correlated Equilibrium," *Games and Economic Behavior* 21, 40–55.

Foster, D. P. and H. P. Young (2003a), "Learning, Hypothesis Testing, and Nash Equilibrium," *Games and Economic Behavior* 45, 73–96.

- Foster, D. P. and H. P. Young (2003b), "Regret Testing: A Simple Payoff-Based Procedure for Learning Nash Equilibrium," University of Pennsylvania and Johns Hopkins University (mimeo).
- Germano, F. and G. Lugosi (2004), "Global Nash Convergence of Foster and Young's Regret Testing," Universitat Pompeu Fabra (mimeo).
- Hart, S. (2004), "Adaptive Heuristics," The Hebrew University of Jerusalem (mimeo).
- Hart, S. and A. Mas-Colell (2000), "A Simple Adaptive Procedure Leading to Correlated Equilibrium," *Econometrica* 68, 1127–1150.
- Hart, S. and A. Mas-Colell (2003), "Uncoupled Dynamics Do Not Lead to Nash Equilibrium," *American Economic Review* 93, 1830–1836.
- Kakade, S. M. and D. P. Foster (2003), "Deterministic Calibration and Nash Equilibrium," University of Pennsylvania (mimeo).
- Jordan, J. (1993), "Three Problems in Learning Mixed-Strategy Nash Equilibria," Games and Economic Behavior 5, 368–386.
- Young, H. P. (2004), Strategic Learning and Its Limits, Oxford: Oxford University Press.